

# **Time Inconsistent Control and Mean Variance Portfolios with State Dependent Risk Aversion**

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- Recap of DynP.
- General problem formulation.
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- Mean Variance with State Dependent Risk Aversion  
(with Xunyu Zhou)

## Standard problem

We are standing at time  $t = 0$  in state  $X_0 = x_0$ .

$$\max_u E \left[ \int_0^T h(s, X_s, u_s) dt + F(X_T) \right]$$

$$dX_t = \mu(t, X_t, u_t)dt + \sigma(t, X_t, u_t)dW_t$$

For simplicity we assume that

- $X$  is scalar.
- The adapted control  $u_t$  is scalar with no restrictions.

We denote this problem by  $\mathcal{P}$

We restrict ourselves to **feedback controls** of the form

$$u_t = u(t, X_t).$$

# Dynamic Programming

We embed the problem  $\mathcal{P}$  in a family of problems  $\mathcal{P}_{tx}$

$\mathcal{P}_{tx}$  :

$$\max_u E_{t,x} \left[ \int_t^T h(s, X_s, u_s) dt + F(X_T) \right]$$

$$dX_s = \mu(t, X_s, u_s) ds + \sigma(s, X_s, u_s) dW_s,$$

$$X_t = x$$

The original problem corresponds to  $\mathcal{P}_{0,x_0}$ .

**Def:**

For  $\mathcal{P}_{tx}$ , we denote the **optimal value function** by  $V(t, x)$  and the **optimal control law** by  $\hat{u}(s, y)$ .

In principle, the optimal control law for  $\mathcal{P}_{tx}$  should be denoted  $\hat{u}_{t,x}(s, y)$ , but:

# Bellman

We now have the Bellman optimality principle, which says that the family  $\{\mathcal{P}_{t,x}; t \geq 0, x \in R\}$  are **time consistent**.

More precisely: If  $\hat{u}$  is optimal on the time interval  $[t, T]$ , then it is also optimal on the sub-interval  $[s, T]$  for every  $s$  with  $t \leq s \leq T$ .

We also have the Hamilton-Jacobi-Bellman equation

**HJB:**

$$V_t(t, x) + \sup_u \left\{ h(t, x, u) + \mu(t, x, u)V_x(t, x) + \frac{1}{2}\sigma^2(t, x, u)V_{xx}(t, x) \right\} = 0,$$
$$V(T, x) = F(x)$$

# Three Disturbing Examples

## Hyperbolic discounting

$$\max_u E_{t,x} \left[ \int_t^T \varphi(T-t) h(X_s, u_s) dt + F(X_T) \right]$$

## Mean variance utility

$$\max_u E_{t,x} [X_T] - \frac{\gamma}{2} \text{Var}_{t,x} (X_T)$$

## Endogenous habit formation

$$\max_u E_{t,x} [\ln (X_T - x + \beta)]$$

$$dX_t = [rX_t + (\alpha - r)u_t]dt + \sigma u_t dW_t$$

## Moral

- These types of problems are **not** time consistent.
- We cannot use DynP.
- In fact, in these cases it is unclear what we mean by “optimality”.

Possible ways out:

- **Easy way:** Dismiss the problem as being silly.
- **Pre-commitment:** Solve (somehow) the problem  $\mathcal{P}_{0,x_0}$  and ignore the fact that later on, your “optimal” control will no longer be viewed as optimal.
- **Game theory:** Take the time inconsistency seriously. View the problems as a game and look for a Nash equilibrium point.

We use the game theoretic approach.

# Ekeland-Lazrak-Pirvu

Maximize expected utility of investment/consumption with hyperbolic discounting

$$\max_u E_{t,x} \left[ \int_t^T \varphi(T-t) h(X_s, u_s) dt + F(X_T) \right]$$

Portfolio dynamics

$$dX_t = [rX_t + (\mu - r)u_t]dt + \sigma u_t dW_t$$

## Results:

- Very precise problem statement and analysis.
- Verification theorem proved.
- Explicit solution when

$$\varphi(T-t) = \alpha e^{a(T-t)} + \beta e^{-b(T-t)}$$

# Basak-Chabakauri

Mean variance optimal investment.

$$\max_u E_{t,x} [X_T] - \frac{\gamma}{2} \text{Var}_{t,x} (X_T)$$

$$dX_t = [rX_t + (\alpha - r)u_t]dt + \sigma u_t dW_t$$

## Results:

- Very nice explicit solution, including hidden Markov model.
- The formal equilibrium problem is never given a precise definition.
- No verification theorem.
- Relies heavily on “total variance formula”, so the method does not generalize from mean-variance.
- The arguments are not completely precise, but more heuristic.

# Contributions of present paper

## Present paper:

- We study a considerably more general problem than in previous papers.
- We derive a system of PDEs, extending the standard HJB equation from DynP.
- Earlier results included as special cases.
- Precise definition of equilibrium given (inspired by Ekeland *et al*).
- Verification theorem proved (inspired by Ekeland *et al*).
- We prove that for every time inconsistent problem there is an equivalent **consistent** problem with the same optimal strategy.
- Particular cases explicitly solved.

## Our Basic Problem

$$\max_u E_{t,x} \left[ \int_t^T C(t, x, X_s, u_s) ds + F(t, x, X_T) \right] + G(t, x, E_{t,x} [X_T])$$

$$dX_s = \mu(t, X_s, u_s) ds + \sigma(s, X_s, u_s) dW_s,$$

$$X_t = x$$

This can be extended considerably.

For simplicity we will consider the easier problem

$$\max_u E_{t,x} [F(X_T)] + G(E_{t,x} [X_T])$$

# The Game Theoretic Approach

- This is a bit delicate to formalize in continuous time.
- Thus we turn to discrete time, and then go to the limit.

## Discrete Time Equilibrium

- We view the problem as a game with a separate player for each point in time  $n = 0, 1, \dots, T - 1$ .
- Player No.  $n$  can only decide on the control  $u_n$  at time  $n$ .
- Given a control sequence  $\bar{u} = \{\bar{u}_k\}_{k=n}^{T-1}$ , the value function for player No.  $n$  is defined by

$$J_n(x, \bar{u}) = E_{n,x} [F(X_T^{\bar{u}})] + G(E_{n,x} [X_T^{\bar{u}}])$$

- A control law (sequence)  $\hat{u}$  is an **equilibrium strategy** if the following hold for each fixed  $n$ .
  - Assume that all players No  $k$  for  $k = n + 1, \dots, T - 1$  use  $\hat{u}_k(\cdot)$ .
  - Then it is optimal for player No  $n$  to use  $\hat{u}_n(\cdot)$ .
- The equilibrium value function is defined by

$$V_n(x) = J_n(x, \hat{u})$$

# Discrete Time Dynamics

**Given:** A controlled Markov process  $\{X_n : n = 0, 1, \dots, T\}$

**Def:**

- For each  $n$  and each fixed real number  $u \in R$  we have the transition probabilities

$$p_n^u(x, dz) = P(X_{n+1} \in dz | X_n = x, u_n = u)$$

- The operator  $\mathbf{P}^u$  is defined for a function sequence  $\{f_n(x)\}$ , where  $f_n : R \rightarrow R$  by

$$(\mathbf{P}^u f)_n(x) = \int_R f_{n+1}(z) p_n^u(x, dz)$$

$$(\mathbf{P}^u f)_n(x) = E[f_{n+1}(X_{n+1}) | X_n = x, u_n = u]$$

- The “infinitesimal operator”  $\mathbf{A}^u$  is defined by

$$\mathbf{A}^u = \mathbf{P}^u - \mathbf{I}$$

## Important Idea

It turns out that a fundamental role is played by the function sequence  $f_n$  defined by

$$f_n(x) = E_{n,x} [X_T^{\hat{u}}]$$

where  $\hat{u}$  is the equilibrium strategy.

The process  $f_n(X_n)$  is of course a martingale under the equilibrium control  $\hat{u}$  so we have

$$\begin{aligned} \mathbf{A}^{\hat{u}} f_n(x) &= 0, \\ f_T(x) &= x. \end{aligned}$$

## Extending HJB

**Proposition:** The equilibrium value function satisfies the system

$$\sup_u \{ \mathbf{A}^u V_n(x) - \mathbf{A}^u (G \circ f)_n(x) + (\mathbf{H}^u f)_n(x) \} = 0,$$

$$V_T(x) = F(x) + G(x)$$

$$\mathbf{A}^{\hat{u}} f_n(x) = 0,$$

$$f_T(x) = x.$$

$$(\mathbf{H}^u f)_n(x) = G(\mathbf{P}^u f_n(x)) - G(f_n(x)), \quad f_n(x) = E_{n,x} [X_T^{\hat{u}}]$$

Note the fixed point character of the problem.

# Continuous Time

The discrete time results extend immediately to continuous time.

- Now  $X$  is a controlled continuous time Markov process with controlled infinitesimal generator

$$\mathbf{A}^u g(t, x) = \lim_{h \rightarrow 0} \frac{1}{h} \{ E_{t,x} [g(t+h, X_{t+h}^u)] - g(t, x) \}$$

- The extended HJB is now an equation with time step  $[t, t+h]$ .
- Divide the discrete time HJB equations by  $h$  and let  $h \rightarrow 0$ .

## Extended HJB Continuous Time

The extended HJB equation for the equilibrium value function:

$$\sup_u \{ \mathbf{A}^u V(t, x) - \mathbf{A}^u (G \circ f)(t, x) + (\mathbf{H}^u f)(t, x) \} = 0,$$

$$\mathbf{A}^{\hat{u}} f(t, x) = 0,$$

$$V(T, x) = F(x) + G(x)$$

$$f(T, x) = x.$$

$$(\mathbf{H}^u f)(t, x) = \lim_{h \rightarrow 0} \frac{1}{h} \{ G(E_{t,x}[f(t+h, X_{t+h}^u)]) - G(f(t, x)) \}$$

Note the fixed point character of the extended HJB.

## The operator $\mathbf{H}^u$

$$\mathbf{H}^u f(t, x) = \lim_{h \rightarrow 0} \frac{1}{h} \{ G (E_{t,x} [f(t + h, X_{t+h}^u)]) - G (f(t, x)) \}$$

We have, to first order,

$$E_{t,x} [f(t + h, X_{t+h}^u)] = f(t, x) + \mathbf{A}^u f(t, x)h$$

Thus, to first order,

$$\begin{aligned} & G (E_{t,x} [f(t + h, X_{t+h}^u)]) \\ &= G (f(t, x)) + G' (f(t, x)) \cdot \mathbf{A}^u f(t, x)h \end{aligned}$$

Thus

$$\mathbf{H}^u f(t, x) = G' (f(t, x)) \cdot \mathbf{A}^u f(t, x)$$

## Extended HJB Continuous Time

The extended HJB equation for the equilibrium value function:

$$\sup_u \{ \mathbf{A}^u V(t, x) - \mathbf{A}^u (G \circ f)(t, x) + G'(f(t, x)) \cdot \mathbf{A}^u f(t, x) \} = 0,$$

$$\mathbf{A}^{\hat{u}} f(t, x) = 0,$$

$$V(T, x) = F(x) + G(x)$$

$$f(T, x) = x.$$

## Diffusion Case

If  $X$  is a scalar SDE of the form

$$dX_t = \mu(X_t, u_t)dt + \sigma(X_t, u_t)dW_t$$

then the extended HJB takes the form

$$\begin{aligned} \sup_u \left\{ \mathbf{A}^u V(t, x) - \frac{1}{2} \sigma^2(x, u) G''(f(t, x)) f_x^2(t, x) \right\} &= 0, \\ \mathbf{A}^{\hat{u}} f(t, x) &= 0, \\ V(T, x) &= F(x) + G(x) \\ f(T, x) &= x. \end{aligned}$$

## The general case

$$\max_u E_{t,x} \left[ \int_t^T C(x, X_s, u_s) ds + F(x, X_T) \right] + G(x, E_{t,x} [X_T])$$

$$dX_s = \mu(X_s, u_s) ds + \sigma(X_s, u_s) dW_s,$$

$$X_t = x$$

## The general case

$$\sup_{u \in \mathcal{U}} \left\{ \left( \mathbf{A}^u V \right) (t, x) + C(x, x, u) - \int_t^T \left( \mathbf{A}^u c^s \right)_t (x, x) ds + \int_t^T \left( \mathbf{A}^u c^{s, x} \right)_t (x) ds \right. \\ \left. - \left( \mathbf{A}^u f \right) (t, x, x) + \left( \mathbf{A}^u f^x \right) (t, x) - \mathbf{A}^u (G \diamond g) (t, x) + \left( \mathbf{H}^u g \right) (t, x) \right\} = 0,$$

$$\mathbf{A}^{\hat{u}} f^y (t, x) = 0,$$

$$\mathbf{A}^{\hat{u}} g (t, x) = 0,$$

$$\left( \mathbf{A}^{\hat{u}} c^{s, y} \right)_t (x) = 0, \quad 0 \leq t \leq s$$

$$V(T, x) = F(x, x) + G(x, x),$$

$$c_s^{s, y} (x) = C(x, y, \hat{u}_s(x)),$$

$$f(T, x, y) = F(y, x),$$

$$g(T, x) = x.$$

## Optimal for what?

In continuous time, it is not immediately clear how to define an equilibrium strategy. We follow Ekeland *et al.*

- Consider a fixed control law  $\hat{u}$ .
- Fix  $(t, x)$  and a “small” time increment  $h$ .
- Choose an arbitrary real number  $u$ .
- Consider the control law  $\bar{u}_h(t, x)$  defined by

$$\bar{u}_h(s, y) = \begin{cases} \hat{u}(s, y) & \text{for } t + h \leq s \leq T \\ u & \text{for } t \leq s \leq t + h \end{cases}$$

**Def:** The control law  $\hat{u}$  is an **equilibrium control** if

$$\lim_{h \rightarrow 0} \frac{J(t, x, \hat{u}) - J(t, x, \bar{u}_h)}{h} \geq 0$$

for all choices of  $t, x, h, u$ .

# Verification Theorem

**Theorem:** Assume that  $V$ ,  $f$  and  $\hat{u}$  satisfies the extended HJB system. Then  $V$  is the equilibrium value function and  $\hat{u}$  is the equilibrium control.

**Proof:** Harder than the corresponding standard verification theorem.

## Equivalent Standard Problems

- Assume that we **know** the equilibrium strategy  $\hat{u}$ .
- Then we can **compute**  $f$  as

$$f_n(x) = E_{n,x} [X_T^{\hat{u}}]$$

- Now **define** the function  $h(t, x, u)$  by

$$h(t, x, u) = (\mathbf{H}^u f)(t, x) - \mathbf{A}^u (G \circ f)(t, x)$$

The extended HJB takes the form

$$\begin{aligned} \sup_u \{ \mathbf{A}^u V(t, x) + h(t, x, u) \} &= 0, \\ V(T, x) &= F(x) + G(x) \end{aligned}$$

This is the HJB for the **time consistent** problem

$$\max_u E_{t,x} \left[ \int_t^T h(s, X_s, u_s) dt + F(X_T) + G(X_T) \right]$$

# Practical handling of the theory

- Make a parameterized Ansatz for  $V$ .
- Make a parameterized Ansatz for  $f$ .
- Plug everything into the extended HJB system and hope to obtain a system of ODEs for the parameters in the Ansatz.

## Basak's Example (in a simple version)

$$dS_t = \alpha S_t dt + \sigma S_t dW_t,$$

$$dB_t = rB_t dt$$

$X_t$  = portfolio value process

$u$  = amount of money invested in risky asset

### Problem:

$$\max_u E_{t,x} [X_T] - \frac{\gamma}{2} \text{Var}_{t,x} (X_T)$$

$$dX_t = [rX_t + (\alpha - r)u_t]dt + \sigma u_t dW_t$$

This corresponds to our standard problem with

$$F(x) = x - \frac{\gamma}{2}x^2, \quad G(x) = \frac{\gamma}{2}x^2$$

## Extended HJB

$$\begin{aligned} V_t + \sup_u \left\{ [rX_t + (\alpha - r)u]V_x + \frac{1}{2}\sigma^2 u^2 V_{xx} - \frac{\gamma}{2}\sigma^2 u^2 f_x^2 \right\} &= 0 \\ V(T, x) &= x \\ \mathcal{A}^{\hat{u}} f &= 0 \\ f(T, x) &= x \end{aligned}$$

**Ansatz:**

$$\begin{aligned} V(t, x) &= g(t)x + h(t) \\ f(t, x) &= A(t)x + B(t) \end{aligned}$$

## Extended HJB

HJB equation becomes:

$$g_t x + h_t + \sup_u \left\{ [rx + (\alpha - r)u]g(t) - \frac{\gamma}{2}\sigma^2 u^2 A^2 \right\} = 0$$
$$g(T) = 1$$
$$h(T) = 0$$

- Embedded static problem:

$$\max_u \left\{ (\alpha - r)g(t)u - \frac{\gamma}{2}\sigma^2 u^2 A^2 \right\}$$

- Optimal control

$$u = \frac{1}{\gamma} \frac{\alpha - r}{\sigma^2} \frac{g(t)}{A^2}$$

Plug back into HJB.

HJB equation becomes:

$$g_t x + h_t + grx + \frac{1}{2\gamma} \frac{(\alpha - r)^2 g(t)^2}{\sigma^2 A^2} = 0$$
$$g(T) = 1$$
$$h(T) = 0$$

Separation of variables gives us

$$g_t + gr = 0$$
$$g(T) = 1$$

We obtain  $g(t) = e^{r(T-t)}$ .

Furthermore

$$h_t + \frac{1}{2\gamma} \frac{(\alpha - r)^2 e^{2r(T-t)}}{\sigma^2 A^2} = 0$$
$$h(T) = 0$$

We need to solve the PDE for the function  $f$ :

$$\mathcal{A}^{\hat{u}} f(t, x) = 0$$

$$f(T, x) = x$$

The PDE becomes:

$$A_t x + B_t + r x A + \frac{1}{\gamma} \frac{(\alpha - r)^2 e^{r(T-t)}}{\sigma^2 A} = 0$$

$$A(T) = 1$$

$$B(T) = 0$$

Separation of variables gives us

$$A_t + Ar = 0$$

$$A(T) = 1$$

We obtain

$$A(t) = e^{r(T-t)}$$

Separation also gives us

$$\begin{aligned}B_t &= \frac{1}{\gamma} \frac{(\alpha - r)^2}{\sigma^2} \\B(T) &= 0\end{aligned}$$

with solution

$$B(t) = \frac{1}{\gamma} \frac{(\alpha - r)^2}{\sigma^2} (T - t)$$

We go back to the equation for  $h$ :

$$\begin{aligned}h_t + \frac{1}{2\gamma} \frac{(\alpha - r)^2}{\sigma^2} &= 0 \\h(T) &= 0\end{aligned}$$

We obtain

$$h(t) = \frac{1}{2\gamma} \frac{(\alpha - r)^2}{\sigma^2} (T - t)$$

## Result

The equilibrium value function and strategy are given by

$$V(t, x) = e^{r(T-t)}x + \frac{1}{2\gamma} \frac{(\alpha - r)^2}{\sigma^2} (T - t)$$

$$\hat{u}(t, x) = \frac{1}{\gamma} \frac{\alpha - r}{\sigma^2} e^{-r(T-t)}$$

$$f(t, x) = e^{r(T-t)}x + \frac{1}{\gamma} \frac{(\alpha - r)^2}{\sigma^2} (T - t)$$

## Equivalent Standard Problem

The Basak problem has the same optimal control as the **time consistent** problem

$$\max_u E_{t,x} \left[ X_T - \frac{\gamma\sigma^2}{2} \int_t^T e^{2r(T-s)} u_s^2 ds \right]$$

$$dX_t = [rX_t + (\alpha - r)u_t]dt + \sigma u_t dW_t$$

We note in passing that

$$\sigma^2 u_t^2 dt = d\langle X \rangle_t$$

# A Closer Look at Mean Variance (with Xunyu Zhou)

We recall that for the Basak problem we have

$$\hat{u}(t, x) = \frac{1}{\gamma} \frac{\alpha - r}{\sigma^2} e^{-r(T-t)}$$

Is this economically meaningful?

**NO!**

Why?

## A Closer Look at Naive Mean Variance

We recall that for the Basak problem we have

$$\hat{u}(t, x) = \frac{1}{\gamma} \frac{\alpha - r}{\sigma^2} e^{-r(T-t)}$$

Is this economically meaningful?

**NO!**

- The control  $u$  is the number of dollars invested in the risky asset. In the Basak case this is independent of the level of wealth.
- You thus invest the same number of dollars in the risky asset regardless of whether your wealth is 100 dollars or 10 billion dollars.
- This is ridiculous.

## Realistic Mean Variance

### Idea:

We let the risk aversion coefficient  $\gamma$  depend on current wealth.

$$\max_u E_{t,x} [X_T] - \frac{\gamma(x)}{2} \text{Var}_{t,x} (X_T)$$

$$dX_t = [rX_t + (\alpha - r)u_t]dt + \sigma u_t dW_t$$

This is covered by the general theory

With  $\beta = \alpha - r$ , the extended HJB system takes the form.

$$\begin{aligned}
& V_t + \sup_{u \in R} \left\{ (rx + \beta u) \left[ V_x - f_y - \frac{\gamma'(x)}{2} g^2 \right] \right. \\
& \left. + \frac{1}{2} \sigma^2 u^2 \left[ V_{xx} - f_{yy} - 2f_{xy} - \frac{\gamma''(x)}{2} g^2 - 2\gamma'(x) g g_x - \gamma(x) g_x^2 \right] \right\} = 0, \\
& f_t(t, x, y) + (rx + \beta \hat{u}) f_x(t, x, y) + \frac{1}{2} \sigma^2 \hat{u}^2 f_{xx}(t, x, y) = 0, \\
& g_t(t, x) + (rx + \beta \hat{u}) g_x(t, x) + \frac{1}{2} \sigma^2 \hat{u}^2 g_{xx}(t, x) = 0,
\end{aligned}$$

with probabilistic interpretations

$$\begin{aligned}
V(t, x) &= E_{t,x} \left[ X_T^{\hat{u}} \right] - \frac{\gamma(x)}{2} \text{Var}_{t,x} \left[ X_T^{\hat{u}} \right], \\
f(t, x, y) &= E_{t,x} \left[ X_T^{\hat{u}} \right] - \frac{\gamma(y)}{2} E_{t,x} \left[ \left( X_T^{\hat{u}} \right)^2 \right], \\
g(t, x) &= E_{t,x} \left[ X_T^{\hat{u}} \right].
\end{aligned}$$

## General Solution

The equilibrium control is given by

$$\hat{u}(t, x) = -\frac{\beta f_x(t, x, x) + \gamma(x)g(t, x)g_x(t, x)}{\sigma^2 f_{xx}(t, x, x) + \gamma(x)g(t, x)g_{xx}(t, x)}.$$

The functions  $f$  and  $g$  are determined by the system

$$f_t(t, x, y) + (rx + \beta\hat{u}) f_x(t, x, y) + \frac{1}{2}\sigma^2\hat{u}^2 f_{xx}(t, x, y) = 0,$$

$$g_t(t, x) + (rx + \beta\hat{u}) g_x(t, x) + \frac{1}{2}\sigma^2\hat{u}^2 g_{xx}(t, x) = 0,$$

with boundary conditions

$$f(T, x, y) = x - \frac{\gamma(y)}{2}x^2,$$

$$g(T, x) = x.$$

The equilibrium value function  $V$  is given by

$$V(t, x) = f(t, x, x) + \frac{\gamma(x)}{2}g^2(t, x).$$

## The Case $\gamma(x) = 1/x$

The equilibrium control is given by

$$\hat{u}(t, x) = \frac{\beta a(t) + \gamma [a^2(t) - b(t)]}{\gamma \sigma^2 b(t)} x$$

where  $a$  and  $b$  solves the ODE system

$$\dot{a} + \left( r + \frac{\beta^2}{\gamma \sigma^2 b} (a + \gamma [a^2 - b]) \right) a = 0,$$

$$a(T, x) = 1,$$

$$\dot{b} + \left\{ 2 \left( r + \frac{\beta^2}{\gamma \sigma^2 b} (a + \gamma [a^2 - b]) \right) + \frac{\beta^2}{\gamma^2 \sigma^2 b^2} (a + \gamma [a^2 - b])^2 \right\} b = 0$$

$$b(T, x) = 1$$

## The Case $\gamma(x) = 1/x$

The equilibrium control is given by

$$\hat{u}(t, x) = c(t)x$$

where  $c$  solves the integral equation

$$c(t) = \frac{\beta}{\gamma\sigma^2} \left\{ e^{-\int_t^T [r + \beta c(s) + \sigma^2 c^2(s)] ds} + \gamma e^{-\int_t^T \sigma^2 c_s^2 ds} \right\}$$